# Hepatoprotective effects of systemic ER activation

ChIPseq/Epigenome genome - Quantification of reads in diffbound promoters and enhancers

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## This script has two parts:

in Part 1, we use thr generated bed file with 182 diffbound promoters (one TSS per gene; the file after diffbind analysis still has multiple TSS per gene, 184 total). This bedfile is then converted into a SAF file using awk in command line. This SAF file, together with BAM files is used for featurecounts to generate a count matrix with all reads in these peak regions (this is analyzed on UPPMAX). The same is done for the 1,816 enhancer peaks.

in Part 2, we import this count matrix into R again, normalize it, calculate significances and generate the plots.

#### Part 1

```
#The enhancer and promoter (with unique TSS, generated in the chunk above) bed files were used to creat
awk 'OFS="\t" {print $1"."$2"."$3, $1, $2, $3, "."}' in.bed > out.saf
#The saf file was then used to annotate the reads, all bam files (Remove duplicates, blocklist, sorted,
#subread version 2.0.0 was used.
Following featureCounts chunk was used:
featureCounts \
        -g gene_id \
        -s 2 \
        -T 2 \
        -M \
        -F SAF \
        -0 \
        -C \
        -a ${SAF_PATH}/example.saf \
        -o ${OUTPUT_PATH}/example.readCount \
        ${BAM_PATH}/*H3K27ac*MkDup.bam \
        &> ${OUTPUT_PATH}/example.readCount.log
the code was then uploaded to uppmax, features counted, then downloaded, finally the header cleaned up
```

### Part 2

Import the reads in peaks which was created using feature counts. Then, normalize the reads to do the analysis. Perform an Anova to compare significances between the conditions.

```
library(tidyverse)
counts_prom <- read.delim("results/Epigenome_analysis/DAc_promoters_142_H3K27ac.clean.readCount", heade</pre>
  dplyr::select("CK0744_H3K27ac_CD2", "CD6_H3K27ac_S2_R1_001", "CK0745_H3K27ac_CD9",
         "CK0746_H3K27ac_HFD3", "CK0747_H3K27ac_HFD4", "HFD6_H3K27ac_S8_R1_001",
         "CK0748_H3K27ac_DPN2", "CK0749_H3K27ac_DPN3", "DPN6_H3K27ac_S15_R1_001",
         "DIP3_H3K27ac_S9_R1_001", "DIP6_H3K27ac_S10_R1_001", "DIP10_H3K27ac_S11_R1_001",
         "E2_2_H3K27ac_S16_R1_001", "CK0750_H3K27ac_E2_8", "CK0751_H3K27ac_E2_9",
         "PPT1_H3K27ac_S12_R1_001", "PPT2_H3K27ac_S13_R1_001", "PPT3_H3K27ac_S14_R1_001")
names(counts_prom) <- c("CDm2","CDm6", "CDm9","HFDm3","HFDm4","HFDm6", "DPN2","DPN3","DPN6","DIP3","DIP</pre>
colsums_prom <- colSums(counts_prom[,])</pre>
#normalise per depth
counts_prom_norm <- sweep(counts_prom, 2, colsums_prom, FUN = "/")</pre>
counts_prom_norm2 <- counts_prom_norm *10^6</pre>
colSums(counts_prom_norm2[,])
First, for promoters.
## CDm2 CDm6 CDm9 HFDm3 HFDm4 HFDm6 DPN2 DPN3 DPN6 DIP3 DIP6 DIP10 E2 2
## 1e+06 1e+06
## E2_8 E2_9 PPT1 PPT2 PPT3
## 1e+06 1e+06 1e+06 1e+06 1e+06
#Combine the replicates
counts_prom_norm3 <- counts_prom_norm2 %>%
  mutate(avg_CD = rowMeans(counts_prom_norm2[,1:3])) %>%
  mutate(avg_HFD = rowMeans(counts_prom_norm2[,4:6])) %>%
  mutate(avg_DPN = rowMeans(counts_prom_norm2[,7:9])) %>%
  mutate(avg_DIP = rowMeans(counts_prom_norm2[,10:12])) %>%
  mutate(avg_E2 = rowMeans(counts_prom_norm2[,13:15])) %>%
  mutate(avg_PPT = rowMeans(counts_prom_norm2[,16:18])) %>%
  tibble::rownames_to_column("loc_ID")
prom_enh_diffbound <- readRDS("results/Epigenome_analysis/annotated_diffbind_and_genomewide_promoters_e.
Prom_Diffbind_HFDup <- prom_enh_diffbound$promoters_HFDup %>% mutate("loc_ID" = paste0(seqnames, ".",st
Prom_Diffbind_HFDdown <- prom_enh_diffbound$promoters_HFDdown %>% mutate("loc_ID" = paste0(seqnames, ".
counts_prom_HFDup <- counts_prom_norm3 %>% filter(loc_ID%in%Prom_Diffbind_HFDup$loc_ID) %>% dplyr::sele
counts prom HFDup2 <- counts prom HFDup %>%
  pivot_longer(cols=2:7) %>% mutate("updown"= "HFDup") %>% group_by(name)
# STATISTICS - One-sided anova.
```

```
counts_prom_HFDup.stat <- aov(value ~ name, data = counts_prom_HFDup2)</pre>
summary(counts_prom_HFDup.stat)
##
                \mathsf{Df}
                      Sum Sq
                                Mean Sq F value
                                                   Pr(>F)
## name
                 5 9.863e+08 197255871
                                         11.99 7.45e-11 ***
## Residuals
               414 6.812e+09 16453050
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
counts_prom_HFDup.stat1 <- TukeyHSD(counts_prom_HFDup.stat)</pre>
counts_prom_HFDup.stat2 <- counts_prom_HFDup.stat1$name</pre>
counts_prom_HFDdown <- counts_prom_norm3 %>% filter(loc_ID%in%Prom_Diffbind_HFDdown$loc_ID) %>% dplyr::
counts_prom_HFDdown2 <- counts_prom_HFDdown %>%
 pivot_longer(cols=2:7) %>% mutate("updown"="HFDdown") %>% group_by(name)
# STATISTICS - One-sided anova.
counts_prom_HFDdown.stat <- aov(value ~ name, data = counts_prom_HFDdown2)</pre>
summary(counts_prom_HFDdown.stat)
##
                Df
                       Sum Sq
                                Mean Sq F value
                                                  Pr(>F)
## name
                 5 9.589e+08 191776541
                                         8.798 5.86e-08 ***
## Residuals 426 9.286e+09 21797457
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
counts_prom_HFDdown.stat1 <- TukeyHSD(counts_prom_HFDdown.stat)</pre>
counts_prom_HFDdown.stat2 <- counts_prom_HFDdown.stat1$name</pre>
counts_prom_HFD_plot <- rbind(counts_prom_HFDup2, counts_prom_HFDdown2)</pre>
counts_prom_HFD_plot$value <- counts_prom_HFD_plot$value+1</pre>
order <- c("avg_CD", "avg_HFD", "avg_DPN", "avg_DIP", "avg_E2", "avg_PPT")
counts_prom_HFD_plot$updown <- factor(counts_prom_HFD_plot$updown, levels = c("HFDup", "HFDdown"))</pre>
ggplot(counts_prom_HFD_plot, aes(x=factor(name, levels=order), y=log2(value), fill=factor(name, levels=
```

ylab("H3K27ac signal\n log2(normalized read count in \nDAc promoter peaks in CDvsHFD)") +

scale\_fill\_manual(values = c("#88CCEE", "#332288", "#DDCC77", "#CC6677", "#AA4499", "#882255"))

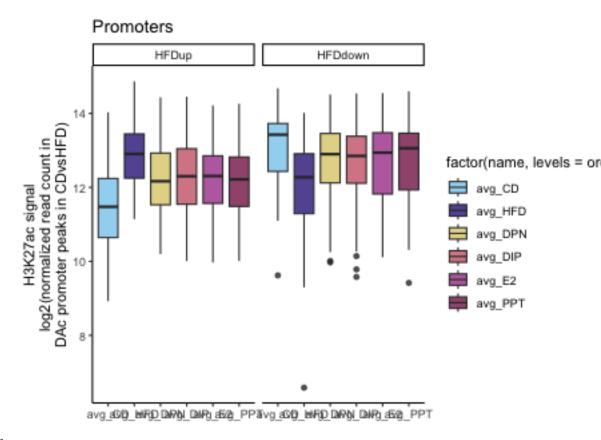
geom\_boxplot(alpha=0.8) +

facet\_wrap(vars(updown)) +

ggtitle("Promoters") +

theme classic() +

xlab("") +

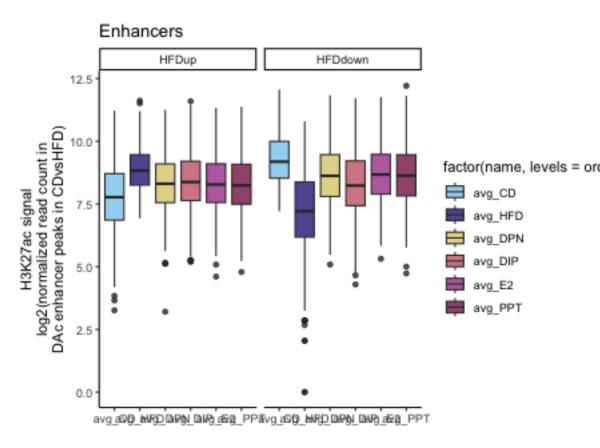


#### Plot for promoters

#### Then, for enhancers

```
## CDm2 CDm6 CDm9 HFDm3 HFDm4 HFDm6 DPN2 DPN3 DPN6 DIP3 DIP6 DIP10 E2_2
## 1e+06 1e+
```

```
#Combine the replicates
counts_enha_norm3 <- counts_enha_norm2 %>%
  mutate(avg_CD = rowMeans(counts_enha_norm2[,1:3])) %>%
  mutate(avg_HFD = rowMeans(counts_enha_norm2[,4:6])) %>%
  mutate(avg_DPN = rowMeans(counts_enha_norm2[,7:9])) %>%
  mutate(avg_DIP = rowMeans(counts_enha_norm2[,10:12])) %>%
  mutate(avg_E2 = rowMeans(counts_enha_norm2[,13:15])) %>%
  mutate(avg_PPT = rowMeans(counts_enha_norm2[,15:18])) %>%
  tibble::rownames_to_column("loc_ID")
Enha_Diffbind_HFDup <- prom_enh_diffbound$enhancers_HFDup %>% mutate("loc_ID" = paste0(seqnames, ".",st
Enha_Diffbind_HFDdown <- prom_enh_diffbound$enhancers_HFDdown %>% mutate("loc_ID" = paste0(seqnames, ".
counts_enha_HFDup <- counts_enha_norm3 %>% filter(loc_ID%in%Enha_Diffbind_HFDup$loc_ID) %>% dplyr::sele
counts_enha_HFDup2 <- counts_enha_HFDup %>%
  pivot_longer(cols=2:7) %>% mutate("updown"= "HFDup")
# STATISTICS - One-sided anova.
res.aov.enha.HFDup.stat <- aov(value ~ name, data = counts_enha_HFDup2)
summary(res.aov.enha.HFDup.stat)
##
                       Sum Sq Mean Sq F value Pr(>F)
                 5 47811754 9562351
                                       86.76 <2e-16 ***
## name
## Residuals 8994 991236041 110211
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
res.aov.enha.HFDup.stat1 <- TukeyHSD(res.aov.enha.HFDup.stat)</pre>
res.aov.enha.HFDup.stat2 <- res.aov.enha.HFDup.stat1$name</pre>
counts_enha_HFDdown <- counts_enha_norm3 %>% filter(loc_ID%in%Enha_Diffbind_HFDdown$loc_ID) %>% dplyr::
counts_enha_HFDdown2 <- counts_enha_HFDdown %>%
 pivot_longer(cols=2:7) %>% mutate("updown"="HFDdown")
# STATISTICS - One-sided anova.
res.aov.enha.HFDdown.stat <- aov(value ~ name, data = counts_enha_HFDdown2)
summary(res.aov.enha.HFDdown.stat)
##
                       Sum Sq Mean Sq F value Pr(>F)
                 5 105312233 21062447
                                        86.72 <2e-16 ***
## name
## Residuals 4080 990931503
                                242875
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
res.aov.enha.HFDdown.stat1 <- TukeyHSD(res.aov.enha.HFDdown.stat)
res.aov.enha.HFDdown.stat2 <- res.aov.enha.HFDdown.stat1$name
counts_enha_HFD_plot <- rbind(counts_enha_HFDup2, counts_enha_HFDdown2)</pre>
counts enha HFD plot$value <- counts enha HFD plot$value+1
```



#### Plot for enhancers

#### sessionInfo()

```
## R version 4.2.3 (2023-03-15)
## Platform: x86_64-apple-darwin17.0 (64-bit)
## Running under: macOS Big Sur ... 10.16
##
## Matrix products: default
## BLAS: /Library/Frameworks/R.framework/Versions/4.2/Resources/lib/libRblas.0.dylib
## LAPACK: /Library/Frameworks/R.framework/Versions/4.2/Resources/lib/libRlapack.dylib
##
## locale:
## [1] en_US.UTF-8/en_US.UTF-8/en_US.UTF-8/C/en_US.UTF-8/en_US.UTF-8
```

```
##
## attached base packages:
## [1] stats
                graphics grDevices utils
                                               datasets methods
                                                                   base
##
## other attached packages:
                                                        dplyr_1.1.2
  [1] lubridate_1.9.2 forcats_1.0.0
                                        stringr_1.5.0
  [5] purrr_1.0.2
                        readr 2.1.4
                                        tidyr_1.3.0
                                                        tibble_3.2.1
## [9] ggplot2_3.4.3
                       tidyverse_2.0.0
##
## loaded via a namespace (and not attached):
## [1] highr_0.10
                         pillar_1.9.0
                                            compiler_4.2.3
                                                              tools_4.2.3
## [5] digest_0.6.33
                         timechange_0.2.0 evaluate_0.21
                                                              lifecycle_1.0.3
## [9] gtable_0.3.3
                          pkgconfig_2.0.3
                                            rlang_1.1.1
                                                              cli_3.6.1
                                                              fastmap_1.1.1
## [13] rstudioapi_0.15.0 yaml_2.3.7
                                            xfun_0.39
## [17] withr_2.5.0
                         knitr_1.43
                                            generics_0.1.3
                                                              vctrs_0.6.3
## [21] hms_1.1.3
                          grid_4.2.3
                                            tidyselect_1.2.0
                                                              glue_1.6.2
## [25] R6_2.5.1
                         fansi_1.0.4
                                            rmarkdown_2.23
                                                              farver_2.1.1
                                                              htmltools_0.5.5
## [29] tzdb 0.4.0
                         magrittr_2.0.3
                                            scales 1.2.1
## [33] colorspace_2.1-0 labeling_0.4.2
                                            utf8_1.2.3
                                                              stringi_1.7.12
## [37] munsell_0.5.0
```